

## 5. FORECASTING: APPLICATIONS

**“Why has U.S. inflation become harder to forecast?”**

**By Stock, J and M., Watson**

## 1 “Why has U.S. inflation become harder to forecast?”

- The rate of price inflation in the United States has become both harder and easier to forecast.
- Easier: inflation is much less volatile than it was in the 1970s and the root mean squared error of inflation forecasts has declined sharply since the mid-1980s.
- Harder: standard multivariate forecasting models do not a better job than simple naive models. The point was made by Atkeson and Ohanian (2001) (henceforth, AO), who found that, since 1984 in the U.S., backwards-looking Phillips curve forecasts have been inferior to a forecast of twelve-month inflation by its average rate over the previous twelve months (naive or random walk forecast).
- Relevance of the topic. Change in terms of forecasting properties can signal changes in the structure of the economy. This can be taken as evidence that suggests that some relations have changed
- What relations? Structural models should be employed (next part of the course).

## 1.1 U.S. Inflation forecasts: facts and puzzles

### 1.1.1 Data

- GDP price index inflation ( $\pi$ ).
- Robustness analysis done using personal consumption expenditure deflator for core items (PCEcore), the personal consumption expenditure deflator for all items (PCE-all), and the consumer price index (CPI, the official CPI-U).
- Real activity variables: the unemployment rate ( $u$ ), log real GDP ( $y$ ), the capacity utilization rate, building permits, and the Chicago Fed National Activity Index (CFNAI)
- Quarterly data. Quarterly values for monthly series are averages of the three months in the quarter.
- The full sample is from 1960:I through 2004:IV.

### 1.1.2 Forecasting models

- Two univariate models and one multivariate forecasting models.
- Let  $\pi_t = 400 \log(p_t/p_{t-1})$  where  $p_t$  is the quarterly price index and let the  $h$ -period average inflation be  $\pi_t^h = (1/h) \sum_{i=0}^{h-1} \pi_{t-i}$ . Let  $\pi_{t+h|t}^h$  be the forecast of  $\pi_{t+h}^h$  using information up to date  $t$ .

### 1.1.3 AR( $r$ )

- Forecasts made using a univariate autoregression with  $r$  lags.  $r$  is estimated using the Akaike Information Criterion (AIC).
- Multistep forecasts are computed by the direct method: projecting  $h$ -period ahead inflation on  $r$  lags
- The  $h$ -step ahead AR( $r$ ) forecast was computed using the model

$$\pi_{t+h}^h - \pi_t = \mu^h + \alpha^h(L)\Delta\pi_t + v_t^h \quad (1)$$

where

1.  $\mu^h$  is a constant
  2.  $\alpha^h(L)$  is a polynomial in the lag operator
  3.  $v_t^h$  is the  $h$ -step ahead error term
- The number of lags is chosen according to the Akaike Information Criterion (AIC) meaning that  $r$  is such that

$$AIC = T \log\left(\sum_{t=1}^T \hat{\varepsilon}_t^2\right) + 2r$$

is minimum. An alternative criterion is the Bayesian Information Criterion (BIC)

$$BIC = T \log\left(\sum_{t=1}^T \hat{\varepsilon}_t^2\right) + r \log(p)$$

#### 1.1.4 AO. Atkeson-Ohanian (2001)

AO forecasted the average four-quarter rate of inflation as the average rate of inflation over the previous four quarters. The AO forecast is

$$\pi_{t+h|t}^h = \pi_t^4 = \frac{1}{4}(\pi_t + \pi_{t-1} + \pi_{t-2} + \pi_{t-3})$$

### 1.1.5 Backwards-looking Phillips curve (PC)

The PC forecasts are computed by adding a predictor to (1) to form the autoregressive distributed lag (ADL) specification,

$$\pi_{t+h}^h - \pi_t = \mu^h + \alpha^h(L)\Delta\pi_t + \beta^h xgap_t + \delta^h(L)\Delta x_t + v_t^h \quad (2)$$

where

1.  $\mu^h$  is a constant
  2.  $\alpha^h(L), \delta^h(L)$ , is a polynomial in the lag operator (lag length chosen using AIC)
  3.  $xgap_t$  is the gap variable (deviations from a low pass filter) based on the variable  $x_t$
  4.  $v_t^h$  is the  $h$ -step ahead error term
- The *PC* forecast using  $u_t = xgap_t = x_t$  and  $\Delta u_t = \Delta x_t$  is called *PC - u*.
  - The forecasts *PC - Δu*, *PC - Δy*, *PC - ΔCapUtil*, *PC - ΔPermits*, *PC - CFNAI* omit the gap variable and only include stationary predictors  $\Delta u$ ,  $\Delta y$ ,  $\Delta CapUtil$ ,  $\Delta Permits$ , *CFNAI*.

## 1.2 Out-of-sample methodology

- The forecasts were computed using the pseudo out-of-sample forecast methodology: that is, for a forecast made at date  $t$ , all estimation, lag length selection, etc. was performed using only data available through date  $t$ .
- The forecasts are recursive, so that forecasts at date  $t$  are based on all the data (beginning in 1960:I) through date  $t$ .
- The period 1960-1970 was used for initial parameter estimation. The forecast period 1970:I–2004:IV was split into the two periods 1970:I–1983:IV and 1984:I–2004:IV.

### 1.3 Results

**Table 1**  
**Pseudo Out-of-Sample Forecasting Results for GDP Inflation**

Multivariate forecasting model:  $\pi_{t+h}^h - \pi_t = \mu^h + \alpha^h(B)\Delta\pi_t + \beta^h xgap_t + \delta^h(B)\Delta x_t + u_t^h$

	1970:I – 1983:IV				1984:I – 2004:IV				$\frac{RMSFE_{84-04}^{h=4}}{RMSFE_{70-83}^{h=4}}$
	h=1	h=2	h=4	h=8	h=1	h=2	h=4	h=8	
AR(AIC) RMSFE	1.72	1.75	1.89	2.38	0.78	0.68	0.62	0.73	
<i>Relative MSFEs</i>									
AR(AIC)	1.00	1.00	1.00	1.00	1.00	<b>1.00</b>	1.00	1.00	0.33
AO	1.95	1.57	1.06	1.00	1.22	1.10	<b>0.89</b>	<b>0.84</b>	0.30
PC- $u$	<b>0.85</b>	0.92	0.88	0.61	<b>0.95</b>	1.11	1.48	1.78	0.42
PC- $\Delta u$	0.87	<b>0.87</b>	0.86	0.64	1.06	1.27	1.83	2.21	0.48
PC- $ugap^{1-sided}$	0.88	0.99	0.98	0.87	1.06	1.29	1.84	2.39	0.45
PC- $\Delta y$	0.99	1.06	0.93	0.58	1.05	1.06	1.23	1.53	0.37
PC- $ygap^{1-sided}$	0.94	0.97	0.99	0.78	0.97	0.97	1.25	1.55	0.37
PC-CapUtil	0.85	0.88	<b>0.79</b>	<b>0.55</b>	0.95	1.01	1.35	1.52	0.43
PC- $\Delta$ CapUtil	1.02	1.00	0.87	0.64	1.03	1.10	1.30	1.51	0.40
PC-Permits	0.93	1.02	0.98	0.78	1.08	1.23	1.31	1.52	0.38
PC- $\Delta$ Permits	1.02	1.04	0.99	0.86	1.00	1.00	1.00	1.02	0.33
PC-CFNAI	.	.	.	.	1.11	1.27	1.86	2.25	.

- The RMSFE of forecasts of GDP inflation has declined and the magnitude of this reduction is striking. In this sense inflation has become easier to forecast
  - The relative performance of the Phillips curve forecasts deteriorated substantially from the first period to the second. This deterioration of Phillips curve forecasts is found for all the activity predictors.
  - The AO forecast substantially improves upon the AR(AIC) and Phillips curve forecasts at the four- and eight-quarter horizons in the 1984-2004 period, but not at shorter horizons and not in the first period.
- ⇒ Substantial changes in the univariate inflation process and in the bivariate process of inflation and its activity-based predictors.

**“Unpredictability and Macroeconomic Stability”**  
**By D’Agostino, A., D. Giannone and P. Surico**

## 2 Unpredictability and Macroeconomic Stability

- D'Agostino Giannone and Surico extend the result for inflation to other economic activity variables: the ability to predict several measures of real activity declined remarkably, relative to naive forecasts, since the mid-1980s.
- The fall in the predictive ability is a common feature of many forecasting models including those used by public and private institutions.
- The forecasts for output (and also inflation) of the Federal Reserves Greenbook and the Survey of Professional Forecasters (SPF) are significantly more accurate than a random walk only before 1985. After 1985, in contrast, the hypothesis of equal predictive ability between naive random walk forecasts and the predictions of those institutions is not rejected for all horizons but the current quarter.
- The decline in predictive accuracy is far more pronounced for institutional forecasters and methods based on large information sets than for univariate specifications.
- The fact that larger models are associated with larger historical changes suggests that the main sources of the decline in predictability are the dynamic correlations between variables rather than the autocorrelations of output and inflation.

## 2.1 Data

- Forecasts for nine monthly key macroeconomic series: three price indices, four measures of real activity and two interest rates:
  1. The three nominal variables are Producer Price Index (PPI ), Consumer Price Index (CPI ) and Personal Consumption Expenditure implicit Deflator (PCED).
  2. The four forecasted measures of real activity are Personal Income (PI ), Industrial Production (IP) index, Unemployment Rate (UR), and EMPloyees on non-farm Payrolls (EMP).
  3. the interest rates are 3 month Treasury Bills (TBILL) and 10 year Treasury Bonds (TBOND).
- The data set consists of monthly observations from 1959:1 through 2003:12 on 131 U.S.macroeconomic time series including also the nine variables of interest.

## 2.2 Forecasting models

The model used are the following:

1. A Naive forecast model (N or RW).
2. Univariate AR, where the forecasts are based exclusively on the own past values of the variable of interest.
3. Factor augmented AR forecast (FAAR), in which the univariate models are augmented with common factors extracted from the whole panel of series.
4. Pooling of bivariate forecasts (POOL): for each variable the forecast is defined as the average of 130 forecasts obtained by augmenting the AR model with each of the remaining 130 variables in the data set.

### **2.3 Out-of-sample methodology**

- Pseudo out-of-sample forecasts are calculated for each variable and method over the horizons  $h = 1, 3, 6,$  and 12 months.
- The pseudo out-of-sample forecasting period begins in January 1970 and ends in December 2003. Forecasts constructed at date  $T$  are based on models that are estimated using observations dated  $T$  and earlier.
- Forecast based on rolling samples using, at each point in time, observations over the most recent 10 years.

## 2.4 Results: full sample

Table 1: *Relative Mean Square Forecast Errors - Full Period*

<i>Random Walk (absolute values)</i>									
hor(m)	PPI	CPI	PCED	PI	IP	UR	EMP	TBILL	TBOND
1	0.45	0.11	0.06	45.58	75.84	0.03	9.45	0.31	0.11
3	1.83	0.59	0.32	13.93	46.23	0.14	7.25	1.29	0.47
6	4.40	1.63	0.94	7.72	35.04	0.45	6.66	2.50	0.99
12	11.87	5.02	2.90	5.03	25.30	1.38	5.75	4.74	2.20
<i>Method AR (relative to RW)</i>									
hor(m)	PPI	CPI	PCED	PI	IP	UR	EMP	TBILL	TBOND
1	0.96	0.83***	0.83***	1.22	0.86*	0.91	0.60***	0.98	0.92
3	1.03	0.88*	0.82**	1.09	0.86	0.81*	0.53***	1.10	1.10
6	1.00	0.84	0.82	1.08	0.94	0.88	0.61***	1.05	1.05
12	1.05	0.93	1.00	1.01	0.95	0.97	0.75***	1.20	1.03
<i>Method FAAR (relative to RW)</i>									
hor(m)	PPI	CPI	PCED	PI	IP	UR	EMP	TBILL	TBOND
1	0.94	0.76***	0.78***	1.15	0.74***	0.72***	0.50***	0.93	0.95
3	0.91	0.71***	0.77**	0.93	0.64**	0.58***	0.39***	1.06	1.19
6	0.84	0.60***	0.75	0.90	0.63*	0.55***	0.43***	0.95	1.17
12	0.84	0.60*	0.83	0.94	0.63	0.64*	0.56***	1.05	1.26
<i>Method POOL (relative to RW)</i>									
hor(m)	PPI	CPI	PCED	PI	IP	UR	EMP	TBILL	TBOND
1	0.94	0.80***	0.80***	1.18	0.80**	0.83***	0.56***	0.94	0.91
3	0.96	0.81***	0.78**	1.02	0.76**	0.73**	0.47***	1.08	1.12
6	0.92	0.72**	0.76*	1.00	0.80*	0.76*	0.54***	0.99	1.07
12	0.92	0.73*	0.85	0.93**	0.78**	0.84***	0.65	1.12	1.07

Notes: Asterisks denote model forecasts that are statistically more accurate than the Naive at 1% (\*\*\*), 5% (\*\*) and 10% (\*) significance levels.

- For all prices and most real activity indicators, the forecasts based on large information are significantly more accurate than the Naive forecasts.
- The factor augmented model produces the most accurate predictions.
- Univariate autoregressive forecasts significantly improve on the naive models for EMP at all horizons and for CPI and PCED at one and three month horizons only. As far as interest rates are concerned, no forecasting model performs significantly better than the naive benchmark.

### 3 Results: sub samples - inflation

Table 2: *Relative MSFEs across Sub-Periods - Inflation*

PERIOD I: sub-sample 1971:1 - 1984:12					PERIOD II: sub-sample 1985:1 - 2002:12					CHANGE
<i>Series: Producer Price Index</i>										
hor	RW	AR	FAAR	POOL	hor	RW	AR	FAAR	POOL	Average
1	0.55	1.03	1.01	0.99	1	0.37	0.89*	0.87*	0.88***	7%
3	2.23	1.05	0.85	0.94	3	1.51	1.01	0.98	0.99**	20%
6	5.79	0.95	0.67	0.82**	6	3.31	1.08	1.08	1.07	34%
12	17.95	1.02	0.65	0.84	12	7.12	1.13	1.20	1.09	33%
<i>Series: Consumer Price Index</i>										
hor	RW	AR	FAAR	POOL	hor	RW	AR	FAAR	POOL	Average
1	0.17	0.83***	0.75***	0.78***	1	0.07	0.85*	0.77**	0.83***	5%
3	0.94	0.84*	0.61***	0.74***	3	0.31	0.99	0.93	0.96**	38%
6	2.85	0.78*	0.46***	0.65***	6	0.68	1.04	1.05	0.98*	83%
12	9.43	0.87	0.44***	0.64**	12	1.57	1.22	1.32	1.16	118%
<i>Series: Personal Consumption Expenditure Deflator</i>										
hor	RW	AR	FAAR	POOL	hor	RW	AR	FAAR	POOL	Average
1	0.08	0.73***	0.71***	0.71***	1	0.05	0.96	0.88**	0.93***	9%
3	0.50	0.72***	0.67**	0.68***	3	0.18	1.04	0.98	1.01	29%
6	1.63	0.72**	0.66*	0.66**	6	0.40	1.13	1.05	1.08	48%
12	5.52	0.92	0.75	0.77	12	0.85	1.37	1.27	1.27	59%

Notes: The column 'change' reads the percentage historical decline in predictability averaged across methods (excluding Naive). Asterisks denote model forecasts that are statistically more accurate than the Naive at 1% (\*\*\*), 5% (\*\*) and 10% (\*) significance levels.

- For all lags except the first, result of AO confirmed, deterioration of the forecasting performance of inflation.

### 3.1 Results: sub samples - real activity

Table 3: *Relative MSFEs across Sub-Periods - Real Activity*

PERIOD I: sub-sample 1971:1 - 1984:12					PERIOD II: sub-sample 1985:1 - 2002:12					CHANGE
<i>Series: Real Personal Income</i>										
hor	RW	AR	FAAR	POOL	hor	RW	AR	FAAR	POOL	Average
1	38.54	1.02	0.95	0.98	1	51.09	1.33	1.27	1.30	21%
3	17.15	1.01	0.86	0.94	3	11.41	1.19	1.01	1.12	14%
6	10.41	1.05	0.83	0.96	6	5.62	1.12	1.01	1.05	2%
12	6.92	0.97	0.84	0.87*	12	3.55	1.07	1.09	1.02	3%
<i>Series: Industrial Production</i>										
hor	RW	AR	FAAR	POOL	hor	RW	AR	FAAR	POOL	Average
1	124.01	0.81*	0.65***	0.75**	1	38.14	0.97	0.95	0.92	14%
3	81.48	0.85	0.55**	0.73**	3	18.64	0.92	0.98	0.88	16%
6	61.42	0.94	0.49*	0.76*	6	14.41	0.97	1.11	0.95	34%
12	43.24	0.95	0.43**	0.72**	12	11.27	0.98	1.22	0.97	62%
<i>Series: Unemployment Rate</i>										
hor	RW	AR	FAAR	POOL	hor	RW	AR	FAAR	POOL	Average
1	0.05	0.86	0.63***	0.78**	1	0.02	0.99	0.88*	0.94***	21%
3	0.25	0.79	0.52***	0.69**	3	0.06	0.91	0.79*	0.84**	18%
6	0.80	0.88	0.49***	0.75	6	0.17	0.85	0.75	0.80*	22%
12	2.42	0.99	0.56**	0.82**	12	0.56	0.93	0.90	0.89	41%
<i>Series: Employees on Nonfarm Payrolls</i>										
hor	RW	AR	FAAR	POOL	hor	RW	AR	FAAR	POOL	Average
1	16.37	0.65***	0.51***	0.60***	1	4.04	0.42***	0.45***	0.40***	4%
3	12.39	0.60**	0.41***	0.53***	3	3.23	0.31***	0.34***	0.29***	-1%
6	11.16	0.70**	0.42***	0.60**	6	3.14	0.37**	0.44*	0.36*	-3%
12	9.21	0.82***	0.49***	0.69***	12	3.05	0.58**	0.72	0.56	8%

Notes: see Table 2.

- Little change in the structure of univariate models for real activity.
- The relative MSFEs of FAAR and POOL suggest that important changes have occurred in the relationship between output and other macroeconomic variables.
- The decline in predictability does not seem to extend to the labor market, especially at short horizons. The forecasts of the employees on nonfarm payrolls are associated with the smallest percentage changes across subsamples.

### 3.2 Results: sub samples - interest rates

Table 4: *Relative MSFEs across Sub-Periods - Interest Rates*

PERIOD I: sub-sample 1971:1 - 1984:12					PERIOD II: sub-sample 1985:1 - 2002:12					CHANGE
<i>Series: 3 Months Treasury Bills</i>										
hor	RW	AR	FAAR	POOL	hor	RW	AR	FAAR	POOL	Average
1	0.64	1.00	0.94	0.95	1	0.05	0.84	0.87	0.81***	-10%
3	2.59	1.12	1.05	1.10	3	0.27	0.98	1.16	0.94**	-5%
6	4.63	1.06	0.88	0.98	6	0.83	1.03	1.25	1.01	11%
12	7.63	1.27	0.93	1.14	12	2.47	1.04	1.34	1.06	8%
<i>Series: 10 Years Treasury Bonds</i>										
hor	RW	AR	FAAR	POOL	hor	RW	AR	FAAR	POOL	Average
1	0.17	0.95	0.96	0.94	1	0.07	0.88**	0.92	0.87***	-9%
3	0.68	1.17	1.21	1.18	3	0.31	1.00	1.15	1.02	-11%
6	1.28	1.07	1.12	1.09	6	0.77	1.02	1.23	1.05	3%
12	2.57	1.04	1.12	1.06	12	1.91	1.01	1.42	1.09	7%

Notes: see Table 2.

- In the second sample the AR, FAAR and POOL methods produce more accurate forecasts than the RW at one month horizon.
- Possible interpretation: increased predictability of the FED due to a better communication strategy.

### **3.3 Results: private and institutional forecasters**

- The predictions for output and its deflator from two large forecasters representing the private sector and the policy institutions are considered.
- The survey was introduced by the American Statistical Association and the National Bureau of Economic Research and is currently maintained by the Philadelphia Fed. The SPF refers to quarterly measures and is conducted in the middle of the second month of each quarter (here the median of the individual forecasts is considered)
- The forecasts of the Greenbook are prepared by the Board of Governors at the Federal Reserve for the meetings of the Federal Open Market Committee (FOCM), which takes place roughly every six weeks.
- Four forecast horizons ranging from 1 to 4 quarters.
- The measure of output is Gross National Product (GNP) until 1991 and Gross Domestic Product (GDP) from 1992 onwards.
- The evaluation sample begins in 1975 (prior to this date the Greenbook forecasts were not always available up to the fourth quarter horizon).

### 3.4 Results: private and institutional forecasters - inflation

Table 5: *Relative MSFEs of Institutional Forecasters - Inflation*

<i>FULL SAMPLE: 1975:1 - 1999:4</i>			
hor(q)	RW	Fed's Green Book(GB)/RW	Survey of Professional Forecasters(SPF)/RW
1	0.26	0.35***	0.37***
2	0.79	0.30**	0.36**
3	1.57	0.29*	0.37
4	2.51	0.32	0.46
<i>PERIOD I: sub-sample 1975:1 - 1984:4</i>			
hor(q)	RW	Fed's Green Book(GB)/RW	Survey of Professional Forecasters(SPF)/RW
1	0.54	0.30***	0.27***
2	1.72	0.21**	0.24**
3	3.51	0.21**	0.25*
4	5.69	0.23*	0.32*
<i>PERIOD II: sub-sample 1985:1 - 1999:4</i>			
hor(q)	RW	Fed's Green Book(GB)/RW	Survey of Professional Forecasters(SPF)/RW
1	0.08	0.58**	0.82
2	0.17	0.93	1.15
3	0.28	0.97	1.39
4	0.39	1.18	1.82

Notes: Asterisks denote rejection of the null hypothesis of equal predictive accuracy between each model and the RW at 1% (\*\*\*) , 5% (\*\*) and 10% (\*) significance levels.

### 3.5 Results: private and institutional forecasters - real activity

Table 6: *Relative MSFEs of Institutional Forecasters - Output*

<i>FULL SAMPLE: 1975:1 - 1999:4</i>			
hor(q)	RW	Fed's Green Book(GB)/RW	Survey of Professional Forecasters(SPF)/RW
1	12.59	0.44**	0.51**
2	9.11	0.49**	0.46**
3	7.45	0.48**	0.50***
4	6.49	0.51**	0.51***
<i>PERIOD I: sub-sample 1975:1 - 1984:4</i>			
hor(q)	RW	Fed's Green Book(GB)/RW	Survey of Professional Forecasters(SPF)/RW
1	25.82	0.37**	0.45**
2	19.01	0.44**	0.41**
3	15.39	0.40***	0.45***
4	13.18	0.42***	0.46***
<i>PERIOD II: sub-sample 1985:1 - 1999:4</i>			
hor(q)	RW	Fed's Green Book(GB)/RW	Survey of Professional Forecasters(SPF)/RW
1	3.77	0.73	0.77
2	2.51	0.77	0.70
3	2.15	0.85	0.73
4	2.03	0.89	0.74

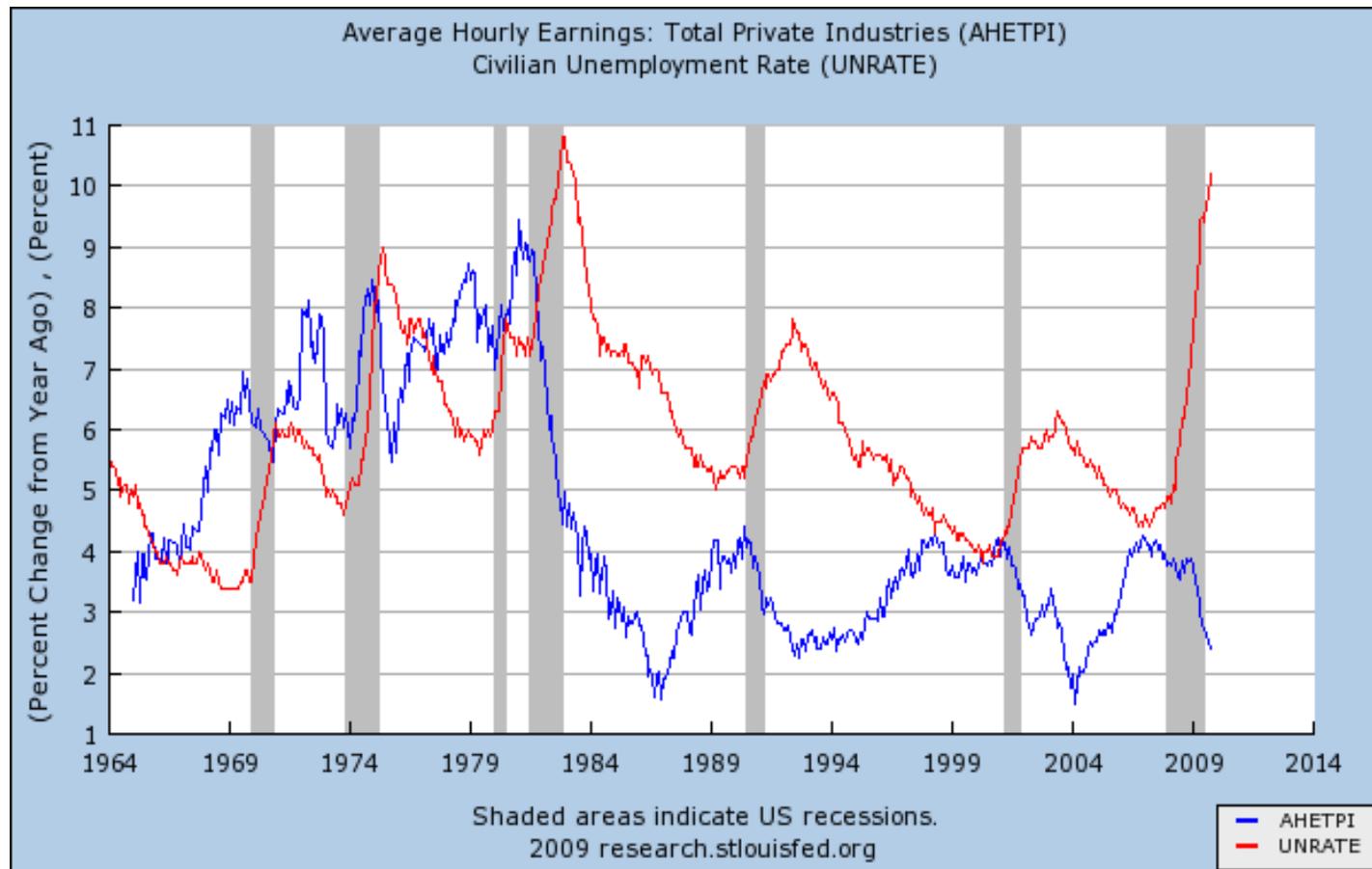
Notes: see Table 5.

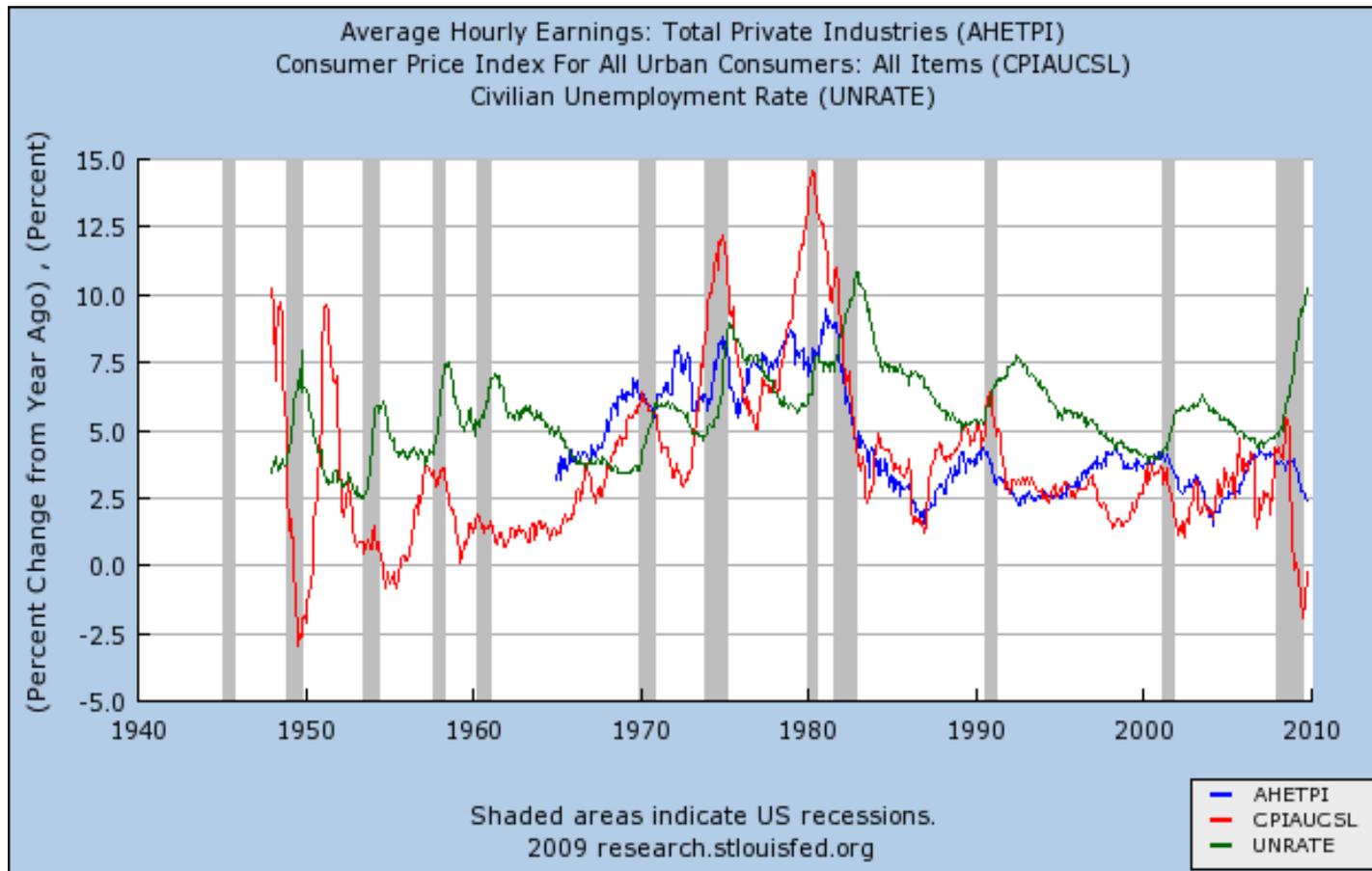
## **”The Return of the Wage Phillips Curve”**

## 4 The Return of the Wage Phillips Curve

- Previous evidence has been taken as a motivation to dismiss the Phillips curve as a theoretical concept.
- Danger with that interpretation.
- In 1958 Phillips uncovered an inverse relation between wage rate inflation and unemployment.
- The focus however in recent years has been shifted to price inflation

## 4.1 Back to the origins





## 4.2 Results

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forecast horizon	RMSE VAR/RW	RMSE AR/RW	% gain using var
1	0.2252	0.2048	9.976
4	0.3642	0.3976	-8.4208
8	0.4892	0.6110	-19.9406
12	0.544	0.6646	-18.1371
16	0.5356	0.6157	-13.0069
18	0.5259	0.5914	-11.0704

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- Phillips curve still (now more than then) characterize dynamics of wage growth and unemployment.
- Crucial question: what has changed in the relation between prices and wages?